

EX SITU CONSERVATION PRIORITIES FOR THE PERUVIAN WILD TOMATO SPECIES (*Solanum* L. SECT. *Lycopersicum* (MILL.) WETTST.)

PRIORIDADES PARA LA CONSERVACIÓN EX SITU DE LAS ESPECIES SILVESTRES DE TOMATE DEL PERÚ (*Solanum* L. SECT. *Lycopersicum* (MILL.) WETTST.)

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Abstract

Tomato (*Solanum lycopersicum*) is a crop of global importance that has center of origin in Peru, with 11 species of wild relatives (CWR) occurring in the country. These CWR contain genetic material that could be used to improve currently cultivated species, and that is usually stored *ex situ* in germplasm banks. Hence, assessing the representativeness of germplasm banks is important to strengthen genetic improvement of tomatoes as well as their resilience to globally changing conditions. Here, we used gap analysis methodology on the 11 species of tomato CWR occurring in Peru. This methodology consists in seven steps that consider representativeness in terms of herbaria/germplasm banks, geography and environments within the country, in order to establish conservation priorities. We also assessed conservation priorities for the Instituto Nacional de Innovación Agraria (INIA), which is the main *ex situ* conservation Institution in Peru. Nationally, we found one species with high priority for conservation: *Solanum huaylasense*, and for INIA we found six species with high conservation priority: *S. huaylasense*, *S. neorickii*, *S. chmielewskii*, *S. corneliomulleri*, *S. arcanum* and *S. chilense*. We also found that the gap analysis methodology allowed for a proper prioritization of species and be readily applied to other species. We conclude by recommending strategies to improve the genetic coverage of the tomato germplasm held at INIA, as well as by discussing priorities for *in situ* conservation of tomato CWRs.

Key words: Gap analysis, agrobiodiversity, conservation priorities, crop wild relatives, *Solanum*.

Resumen

El tomate (*Solanum lycopersicum*) es un cultivo de importancia mundial que tiene un centro de origen en el Perú, con 11 especies de parientes silvestres (CWR) presentes en el país. Estos CWR contienen material genético que podría usarse para mejorar especies actualmente cultivadas, y que generalmente se almacenan *ex situ* en bancos de germoplasma. Por lo tanto, la evaluación de la representatividad de los bancos de germoplasma es importante para fortalecer la mejora genética de los tomates, así como su resistencia a las condiciones mundiales cambiantes. Aquí, utilizamos la metodología de análisis de vacíos en las 11 especies de CWR de tomate presentes Perú. Esta metodología consiste en siete pasos que consideran la representatividad en términos de bancos de herbarios / germoplasma, geografía y ambiental dentro del país, a fin de establecer prioridades de conservación *ex situ*. También evaluamos las prioridades de conservación *ex situ* para el Instituto Nacional de Innovación Agraria (INIA), que es la principal institución de conservación *ex situ* en el Perú. A nivel nacional, encontramos una especie con alta prioridad para conservación: *Solanum huaylasense*, y para INIA encontramos seis especies con alta prioridad de conservación: *S. huaylasense*, *S. neorickii*, *S. chmielewskii*, *S. corneliomulleri*, *S. arcanum* y *S. chilense*. También encontramos que la metodología de análisis de vacíos permitió una priorización adecuada de las especies y se puede aplicar fácilmente a otras especies. Concluimos recomendando estrategias para mejorar la cobertura genética del germoplasma de tomate en el INIA, así como discutiendo las prioridades para la conservación *in situ* de los CWRs del tomate.

Palabras clave: análisis de vacíos, agrobiodiversidad, prioridades de conservación, parientes silvestres de cultivos, *Solanum*.

Introduction

Crop wild relatives (CWR) are wild plant species that share a relatively recent common ancestry with cultivated plants. CWR typically possess wider genetic

diversity in comparison with their cultivated counterparts due to their constant interaction with the environment. That genetic diversity is lost during the domestication process (Castañeda-Alvarez *et al.*,

2015a). However, those genes found in CWRs might be introgressed into the crop by genetic improvers, given the ease of transfer of genes between closely related species (Harlan & de Wet, 1971; Singh, 2001; Castañeda-Alvarez *et al.*, 2015a). CWR have provided crops with traits such as pest and disease resistance, tolerance to abiotic stresses, increased yield, male sterility, and quality, increasing the value and sustainability in as many as 15 crops, many of them of worldwide importance, such as maize, potatoes, or rice. In the past 20 years, there has been a steady increase in the rate of release of cultivars containing genes from CWR, and their contribution should only increase as the development of molecular technologies makes identification and utilization of diverse germplasm more efficient (Dempewolf *et al.*, 2017, Singh, 2001; Hajjar & Hodgkin, 2007; Prescott-Allen & Prescott-Allen, 1986; Tanksley & McCouch, 1997). Plant breeders obtain CWR material from genebanks, hence having a representative sample of such material is of critical importance, especially at origin centers, where their diversity is the highest. Tomate is one of the ten crops with the most breeding uses of CWR documented (Dempewolf *et al.*, 2017). However, major gaps in the genetic diversity of important crop gene pools such as tomato remain to be filled in *ex situ* germplasm collections (FAO, 1997; Maxted & Kell, 2009; Khoury *et al.*, 2010).

A few methodologies have been described to determine gaps in germplasm collections. One of those methodologies is called “gap analysis”, and consists in the determination of a priority score per species with a corresponding mapped distribution of gaps (Ramírez-Villegas *et al.*, 2010). Gap analysis combines sampling, geographic and ecological/environmental representativeness scores to determine the extent of conservation of the wild relatives in *ex situ* conservation institutions. This methodology has been used widely to date (Castañeda-Alvarez *et al.*, 2015a, b; Ramírez-Villegas *et al.*, 2010; Khoury *et al.*, 2010), and is efficient for prioritizing CWR taxa considering wide variation in the potential diversity encompassed in each taxon’s distribution and the lack of molecular diversity data. Based on that methodology, three types of scores are calculated: (1) sampling representativeness, (2) geographic representativeness, and (3) environmental representativeness. The sampling representativeness score allows to determine the total number of germplasm accessions estimated as sufficient to represent a taxon, relative to the known extent of the taxon utilizing all gene bank and reference (herbaria) data regardless of whether geographical coordinates are available. Geographic and environmental scores are used as proxies for genetic diversity and potential functional adaptation to diverse environments, respectively, based on the assumption that the genetic composition of plant species varies across its geographic range and is associated with

adaptation to different ecological conditions (Debouck, 2009). This implies that a more representative collection of germplasm is more likely to capture a greater diversity of alleles adapted to different conditions locally. Together these three indices represent a powerful tool to assess gaps in germplasm collections, and also inform the strategies to fill those gaps as they can directly projected into maps.

Peru is considered a center of origin of tomato species, given that it has 84% of the species considered as tomatoes (Tanksley & McCouch, 1997). Within *Solanum Section Lycopersicum*, 13 species have been reported around the world, with 11 of those occurring in Peru, and three of them endemic (*S. corneliomulleri*, *S. huaylasense* y *S. arcanum*) (Peralta *et al.*, 2008). Additionally, tomato has a great richness in terms of varieties, having diversity in shape, size, color, taste, texture and hardship, among other commercially important characters. Tomato has also been used as a model organism for genetic, physiological, cellular, and biochemical studies (Cocalliadis, 2012). However, the conservation of these species is incipient in Peru, as well as their use in genetic improvement. Moreover, the country imports genetically improved varieties. In Peru, the Instituto Nacional de Innovación Agraria (INIA) is responsible for research in genetic resources, and hence any conservation effort should start from its collections. Currently, INIA has a wild tomato germplasm bank with 156 accessions from seven species, but further collections are needed in the center of origin of such species. Here, we aim to (1) analyze the collection gaps for Peru on wild tomato species, (2) analyze the collection gaps for INIA’s tomato CWR collection, and (3) test the gap analysis methodology. We hypothesize that by combining the known information on the distribution of these species, we will be able to identify new collecting sites, along with their priority. This in turn will allow strengthening *ex situ* and *in situ* conservation efforts for CWR. We conclude our analysis by identifying potential areas for wild tomato *in situ* conservation.

Material and methods

This article follows and adapts the methodology proposed by Ramírez-Villegas *et al.* (2010). Our methodology is composed by a seven-step gap analysis process which is designed to evaluate conservation deficiencies considering the (1) taxonomic, (2) geographic and (3) environmental levels. The aim of this method is to define the extent to which current genebank holdings of Peruvian tomatoes at both national and international institutions represent the total genetic diversity within tomato CWRs. We also use this method with INIA’s tomato CWR germplasm collection. The method averages three scores in order to produce a table presenting the high, medium and low priority species for collecting. Using this prioritization and species distribution models, we can identify

potential collecting areas for high priority species, as well as overlapping high priority regions for the collection of multiple taxa identified and/or high priority sites for *in situ* conservation. The method was conducted as follows:

Determination of target taxa, delineation of target area and harvesting of occurrence data

This stage of the method involves three steps:

a. Identification of the target cultivated species, which in this case is *Solanum lycopersicum*, along with its wild relatives that occur within Peru. We used Peralta *et al.* (2008) taxonomy in order to determine the taxa to study.

b. Creation of a database containing as many records as possible both of genebank accessions and herbarium specimens of both national and international databases, along with (when available) their respective passport data, specifically the names of the places of collection and coordinates (i.e. latitude and longitude). Data was collected from the Global Biodiversity Information Facility (GBIF), from GENESYS (GENESYS, 2018), from Solanaceae Source, and from INIA's database on March 2018.

c. Cross-check, verification, and correction of geographic references (coordinates) through thorough review of data and use of verification tools, Google Earth, and high detail physical maps of localities, and strict selection only of verified geo-referenced samples for distribution modeling, as the quality of location data strongly affects the performance of niche modeling techniques (Graham *et al.*, 2008). The final dataset is available at Sotomayor *et al.* (2019).

Determination of sampling deficiencies at the taxon level

A first estimate of collection priority is calculated for each taxon using the “Sampling Representativeness Score” (SRS, Equation 1), which compares total germplasm accessions to herbarium records.

$$SRS = (GS)/(GS + HS) * 10$$

Equation 1. Calculation of the Sampling Representativeness Score (SRS).

GS: number of germplasm samples; HS: herbarium samples.

SRS is calculated as the number of germplasm samples (GS) divided by the total number of samples, i.e. the sum of germplasm plus herbarium samples (HS), regardless of whether samples contain geographical location coordinates. This score (SRS) allows for a general estimation of adequacy of germplasm holdings of each taxon based upon all available data at different genebanks and herbariums. In the case that a taxon has no genebank samples, it is listed as a “high priority species” for collecting by setting the FPS (see step 7 below) to 0. Mapping of herbarium samples and genebank accessions was performed using Q-GIS (version 3.2.2) (QGIS

Development Team 2018) in order to provide a general geographic assessment of the available data. We calculated two types of SRS, one using all data for the country, and a second one only using germplasm data from INIA's bank.

Create potential distribution models for taxa

Potential distributions of taxa are calculated using the maximum entropy (Maxent) model (Phillips *et al.*, 2006), with a set of bioclimatic variables and species presence data as inputs. This set included 19 bioclimatic variables (BioClim) that were extracted from the website of WorldClim (WorldClim, 2005), (Hijmans *et al.*, 2005), as well as the elevation associated with each geographic coordinate. BioClim includes temperature, rainfall and other variables calculated from their periodicity or range. We used Maxent due to its precision and confidence when predicting species distributions, even with low number of species records, as it has been elsewhere in the literature (Phillips *et al.*, 2006; Hernandez *et al.*, 2006; Loiselle *et al.*, 2008; Costa *et al.*, 2010; Phillips & Dudik, 2008; Dormann, 2006; Elith *et al.*, 2006; Hijmans & Graham, 2006). Default features were used in Maxent, in which complexity of the models (represented by the number of terms and the type of interactions between environmental variables) depend upon the number of input data points (Phillips *et al.*, 2006; Hernandez *et al.*, 2006; Loiselle *et al.*, 2008; Costa *et al.*, 2010; Phillips & Dudik, 2008). Background points for model training equal 10 000 random points over the distributional range of the gene pool (each species) in order to avoid overfitting (VanDerWal *et al.*, 2009; Phillips, 2008). Ten runs were conducted per species in order to determine their potential distribution. We used an arbitrary threshold that would us to capture the top 90% of the predictions of the average model, in order to map the potential distribution coverage (PDC). Finally, we mapped the PDC for each taxa together with their respective collection coordinates in order to display the distribution of each species and to visually determine their correspondence.

Geographic coverage assessment

The adequacy of geographic coverage of accessions per species was calculated from the respective thresholded Maxent model (PDC) in comparison with the geographic coverage of the reported distribution of each species (GCS). This geographic coverage was estimated using a 10 km buffer around each georeferenced point of the reported distribution of each species, after removing duplicates. These values were used to calculate the Geographic Representativeness Score (GRS, Equation 2).

$$GRS = (GCS)/(PDC) * 10$$

Equation 2. Calculation of the Geographic Representativeness Score (GRS).

GCS= geographic coverage of the reported distribution of each species;

PDC= potential distribution coverage of the taxon under analysis.

GRS is thus the geographic coverage of the reported distribution of collections (GCS) divided by the potential distribution coverage of the taxon under analysis (PDC). A higher the GRS, means a higher the representativeness of collections compared to the potential distribution of the taxon.

Environmental coverage assessment

The adequacy of environmental coverage of genebank accessions is calculated as an ‘Environmental Representativeness Score’ (ERS, Equation 3), assessed by comparing the germplasm samples in relation to the full environmental range of the modeled taxon distribution. The same set of climatic layers used for developing the potential distribution coverage (the MaxEnt models) are used to perform a principal components analysis, after standardization. The first two principal components (which accounted for more than 70% of the variability in each species) were reclassified into twenty equal classes.

$$ERS = \sum Wi (ECi/PEDi) * 10$$

Equation 3. Calculation of the Environmental Representativeness Score (ERS), the adequacy of environmental coverage of genebank accessions. EC: environmental coverage of germplasm collections; PED: potential environmental coverage of the taxon under analysis; w: weight of the principal component.

For these two principal components ($i = 2$), ERS is calculated as the environmental coverage (i.e. number of different classes) of germplasm collections (EC) divided by the potential environmental coverage of the taxon under analysis (PED), times the weight (W) of the principal component (weights of the two components are rescaled so that the sum of their weights is 1).

Numeric assessment to determine the priority of collecting for *ex situ* conservation for each taxa

All representativeness scores (SRS, GRS, and ERS) were averaged with equal weight to obtain a final score of prioritization of species (Khoury *et al.*, 2019; Castañeda-Alvarez *et al.*, 2015b; Ramírez-Villegas *et al.*; 2010). The “Final Priority Score (FPS)”, is then used to classify taxa according to the following ranges (Ramírez-Villegas *et al.*, 2010): (1) as high priority species if the FPS is between 0 and 3, (2) as medium priority species if the FPS is between 3.01 and 5, (3) as low priority species if the FPS is between 5.01 and 7.5, and (4) as well conserved species (no need for further collection) if the FPS is between 7.51 and 10. This assessment was conducted both for the entire country scores, as well as for INIA’s scores.

Prioritization of geographic areas for collecting germplasm

The potential collection zones for each high priority species were identified separately in order to determine “collection gaps”. This is done by contrasting potential distributions with collection sites and by combining those collection gaps for the prioritization. These gaps represent zones where further collection missions should be conducted in order to expand the genetic base of germplasm collections and represents the final result of this methodology. We also produced a combined map of all potential distributions of the species analyzed, in order to identify areas of high priority for collection of multiple tomato CWR, as well as areas for the prioritization of *in situ* conservation actions of these species.

Results

We analyzed 11 species of wild relatives that are distributed in Peru (Figure 1). Generally, there were more collections in germplasm banks (Table 1) than in herbaria (Table 2), although, in both cases, the majority of records did not have geographic coordinates. The gap analysis methodology allowed us to successfully assess the *ex situ* conservation priorities for all 11 species (Table 3). One species was listed as high conservation priority, *S. huaylasense* (Figure 2a), at the national level; while for INIA, six species were listed as high priority for collection (Figure 2): *S. huaylasense*, *S. chmielewskii*, *S. corneliomullerii*, *S. neorickii*, *S. chilense*, and *S. arcanum*. *S. huaylasense* only had 3 records from herbaria and one from germplasm banks, so it was listed with the highest priority of all species analyzed. At the national level, five other species were listed as medium priority for collection: *S. neorickii*, *S. chmielewskii*, *S. corneliomulleri*, *S. arcanum*, and *S. chilense*; four species as low priority for collection: *S. pennellii*, *S. lycopersicum* var. *cerasiforme*, *S. habrochaites* and *S. peruvianum*; and one species as well conserved: *S. pimpinellifolium*. For INIA, five species were listed as medium priority (Figure 3): *S. pimpinellifolium*, *S. peruvianum*, *S. penellii*, *S. lycopersicum* var. *cerasiforme*, and *S. habrochaites*. All three endemic species to Peru (*S. corneliomulleri*, *S. huaylasense*, *S. arcanum*) were listed as high priority for INIA collection, while only *S. huaylasense* was listed as high priority at the national level, and the other two endemics as medium priority. Two main areas had a major concentration of tomato CWRs, one in the north towards La Libertad Region, and one in the south towards Arequipa Region (Figure 4).

Discussion

Here we aimed to conduct a gap analysis of the Peruvian crop wild relatives (CWR) of the tomato and to test the aforementioned methodology. We successfully assessed the conservation priorities for all

11 CWR species from Peru and found contrasting priorities among species both at the national level, as well as for INIA's collection. Firstly, at the national level, we found that one of three endemic species to Peru (*S. huaylasense*) had a high conservation priority, while the other two endemics (*S. arcanum* and *S. corneliomulleri*) had medium conservation priority, along with three other species. These results contrasted with those found at INIA, where all three endemics plus three other species had high conservation priorities; and the other five species analyzed had medium conservation priorities. In both assessments, only one species, *S. pimpinellifolium* at the national level, had no urgency of collection and hence is well represented in current germplasm banks. This last species corresponds to a species that widely distributed in the country. Secondly, we conducted an adapted version of gap analysis (Ramírez-Villegas *et al.*, 2010) in which we were able to estimate three different scores in order to construct a final priority score. Overall, this gap analysis on tomato CWRs represents an important outcome for the conservation of CWR at a center of origin and could serve as a starting point for other gap analyses for species that have their center of origin in Peru. This also represents the first analysis of *ex situ* conservation priorities for this crop, and its recommendations should be taken into consideration for future efforts on capturing the genetic diversity of tomato's CWRs.

The second aim of this article was to test the gap analysis methodology (Maxted *et al.*, 2008; Ramírez-Villegas *et al.*, 2010), and if needed to adapt it to the available data. We were able to follow the steps of calculating taxonomic gaps in the collections and running species distributions models to estimate geographic and environmental gaps. However, we were not able to conduct an analysis on rarity due to the lack of good representativeness among widely collected species along the environmental axes (Ramírez-Villegas *et al.*, 2010). To replace that criterion, we can propose the addition of a consideration on endemism, with the purpose of assigning extra priority to species that endemics, or highly restricted endemics. This criterion is important, especially in a country that is also a center of origin and that has many endemic species. Overall, this adapted methodology was useful and effective in determining gaps for prioritization of *ex situ* conservation collection, both at the national and INIA's level. More analysis is needed to deal with environmental/ecological representativeness using multivariate statistics, as well as to analyze rarity. These refinements can be implemented in a crop that has more accessions, or on a gap analysis for entire distributions of CWR. This study represents a well-documented study case, and hopefully can be used by other practitioners to conduct gap analyses on the CWR of other crops.

Our gap analysis allowed us to determine conservation priorities for the Peruvian CWR of the tomato. Given these results, the next step is to incorporate these results in the design of a strategy to enrich the current tomato genebank at INIA. This strategy should focus on: (1) collection missions for species with high and medium conservation priority, by evaluating the maps generated with Maxent in contrast with current collections to estimate collection gaps, and then, to super-impose collection gaps between species; (2) collection missions to complete specific gaps for species with low priority also by comparing Maxent models and geographic coordinates of current collections, and if possible relate these gaps to the collection gaps of the high/medium priority species; and (3) assessment of the genetic diversity of species that do not have any priority for collection, in order to determine the degree of duplication among samples. Together, these lines of action will allow for the conservation of a comprehensive sample of tomato's genepool in Peru and can also serve to implement strategies in other species with enough data for gap analysis.

One of the products of gap analyses are maps highlighting sites where the species under study are more likely to occur (Castañeda-Alvarez *et al.*, 2015a, b; Ramírez-Villegas *et al.*, 2010). These maps guide future collection efforts but can also be used to prioritize sites for *in situ* conservation of such species, especially after recognizing sites where multiple CWRs occur. These maps of species distributions can be used as inputs for systematic conservation planning (Margules & Pressey, 2000) in order to identify sites where *in situ* conservation of these species can be achieved. As an example of this process, in Figure 4 we combined the distributions of tomato CWR species with the Peruvian National System of Protected Areas and found that the tomato CWR hotspot towards the northern part of the country (La Libertad Region) coincides with two protected areas Calipuy and Sunchubamba. A recommendation arising from this study could be to actively track the tomato CWR species in those protected areas. Alternatively, a national protected area (e.g. an agricultural biodiversity zone) could be created to conserve tomato CWRs in the southern hotspot. Hence, gap analyses can also be used to delineate *in situ* conservation strategies, as they have been used for a global analysis to identify sites at the global scale for CWR *in situ* conservation (Vincent *et al.*, 2019).

Conclusions

Gap analysis has proved to be an effective and useful methodology for the determination of *ex situ* conservation priorities. We successfully assessed the conservation priorities for all 11 CWR species from Peru and found that one of three endemic species to Peru (*S. huaylasense*) had a high *ex situ* conservation

priority, while the other two endemics (*S. arcanum* and *S. corneliomulleri*) had medium *ex situ* conservation priority. Additionally, this methodology can also guide the prioritization of sites for *in situ* conservation, as we identified two hotspots of tomato CWRs diversity. Next steps to optimize this methodology could be to interact with experts in tomato within the country or abroad in order to validate the priorities found here. Further refinements could also be in the direction of improving Maxent models for very restricted endemic species (e.g. *S. huaylasense*) that yielded models overpredicting their distribution. Alternatively, other methodologies can be developed to deal with such species. Genetic representativeness of current germplasm collections can also be another line of action to complement gap analysis and determine the representativeness of a collection. Overall, this gap analysis methodology represents an opportunity to develop further analyses on other species and in countries that are centers of origin of widely cultivated species.

Acknowledgements

Funding was provided by the Project “Discovering the potential for genetic improvement of tomato quality in the wild tomato germplasm of Peru”, funded by the Peruvian National Program of Agricultural Innovation, Project #103.

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Apéndice.

Tablas y Figuras.

Table 1. List of Genebank Institutions from which data on wild tomato species was collected, including the number of records obtained.

Institution	Number of records with coordinates	Number of records without coordinates
Bioversity International	104	1 099
Centro Agronómico Tropical de Investigación y Enseñanza (CATIE).	0	174
International Plant Genetic Resources Institute (IPGRI) (ITA303).	1	0
Nordic Genetic Resources (NORDGEN)	0	10
Polish gene bank passport data of plants accessions which are important in human life (POL030).	0	14
The System-wide Information Network for Genetic Resources (SINGER)	43	332
United States Department of Agriculture Agricultural Research Service _National Plant Germplasm System (USDA_NPGS)	404	2 936
CZE National PGR Inventory	0	1
Tomato Genetic Resources Center, Department of Vegetable Crops, University of California (USA176).	403	2 768
Department of vegetable crops university of California Scientific Center of Agrobiotechnology (ARM008).	48	643
Australian Grains Genebank, Department of Economic Development Jobs Transport and Resources (AUS165).	0	120
Plant Production Research Center Piestany (SVK001).	0	2
External Branch North of the Department Genebank, IPK, Potato Collection in Gross-Luesewitz (DEU159).	0	2
Genebank UPV (ESP026).	18	107
Gobierno de Aragón. Centro de Investigación y Tecnología Agroalimentaria. Banco de Germoplasma de Hortícolas (ESP027).	0	1
Consejo Superior de Investigaciones Científicas. Estación Experimental La Mayora (ESP058)	1	75
Botanical Garden, University of Nijmegen (NLD020)	0	8
Centre for Genetic Resources, the Netherlands Plant Research International (NLD037).	30	33
Plant Breeding and Acclimatization Institute (POL003).	0	9
World Vegetable Center (TWN001).	0	768
Northeast Regional Plant Introduction Station, Plant Genetic Resources Unit, USDA-ARS, New York State Agricultural Experiment Station, Cornell University (USA003)	95	348
National Seed Storage Laboratory, USDA-ARS (USA005)	0	4
National Center for Genetic Resources Preservation (NCGRP) (USA995)	0	4
Total	1 147	9 459

Table 2. List of Herbarium Institutions from which data on wild tomato species was collected, including the number of records obtained.

Institution	Number of records with coordinates	Number of records without coordinates
Herbarium Berolinense	0	1
Bioversity International	549	284
Consortium of Pacific Northwest Herbaria (CPNWH)	0	2
Royal Botanic Garden Edinburgh Herbarium (E)	0	1
European Bioinformatics Institute (EMBL-EBI)	1	0
Field Museum of Natural History (Botany) Seed Plant Collection	18	29
Fairchild Tropical Botanic Garden Virtual Herbarium Darwin Core format	1	0
Herbarium arequipense (HUSA)	15	8
Vascular plants collection (P) of the Museum national d'Histoire Naturelle (MNHN - Paris)	0	1
Missouri Botanical Garden	9	21
Herbarium of The New York Botanical Garden	85	49
Phanerogamic Botanical Collections (S)	0	3
Database Schema for UC Davis	0	310
University of British Columbia	0	4
United States Department of Agriculture Agricultural Research Service _National Plant Germplasm System USDA_NPGS)	17	205
VIT Herbarium - Vascular Plants (The Natural History Museum of Alava)	2	0
Total	697	918

Table 3. Number of accessions analyzed per species, along with their respective gap analysis scores. Data presented includes Herbarium samples (HS) and Germplasm samples (GS), along with their respective geographically unique points (UHS and UGS). Gap analysis scores include the following: sampling representativeness score (SRS) for both INIA and Peru, geographic representativeness score (GRS), the environmental representativeness score (ERS), and the final priority score (FPS) for all species included in the analysis, for both INIA and Peru. Species with high collection priority are bolded.

Taxon	HS (UHS)	GS (UGS)	Total	SRS (INIA)	SRS (PERU)	GRS	ERS	FPS (INIA)	FPS (PERU)
<i>Solanum pimpinellifolium</i>	475 (101)	2 495 (216)	2 970 (317)	0.250	8.401	9.264	5.182	4.898	7.615
	448	1 388	1 836						
<i>S. peruvianum</i>	(85)	(122)	(207)	0.000	7.560	6.672	7.224	4.632	7.152
	110	519	629						
<i>S. pennellii</i>	(26)	(92)	(118)	0.016	8.251	4.274	4.945	3.079	5.823
	35	88	123						
<i>S. neorickii</i>	(15)	(16)	(31)	0.081	7.154	2.031	3.724	1.945	4.303
<i>S. lycopersicum</i> var. <i>cerasiforme</i>	113	694	807						
	(19)	(86)	(105)	0.583	8.600	1.722	8.371	3.559	6.231
	9	6	15						
<i>S. huaylasense</i>	(3)	(1)	(4)	0.000	4.000	0.244	--	0.244	1.415
	125	482	607						
<i>S. habrochaites</i>	(50)	(53)	(103)	0.114	7.941	4.336	6.803	3.751	6.360
	12	270	282						
<i>S. chmielewskii</i>	(0)	(19)	(19)	0.070	9.574	0.997	--	1.068	3.524
	61	76	137						
<i>S. corneliomulleri</i>	(15)	(9)	(24)	0.072	5.547	1.452	3.171	1.565	3.390
	87	45	132						
<i>S. arcanum</i>	(22)	(11)	(33)	0.000	3.409	2.136	4.184	2.107	3.243
	43	247	290						
<i>S. chilense</i>	(11)	(16)	(27)	0.000	8.517	1.038	5.167	2.068	4.907

PERUVIAN TOMATO GAP ANALYSIS

Julio - Diciembre 2019

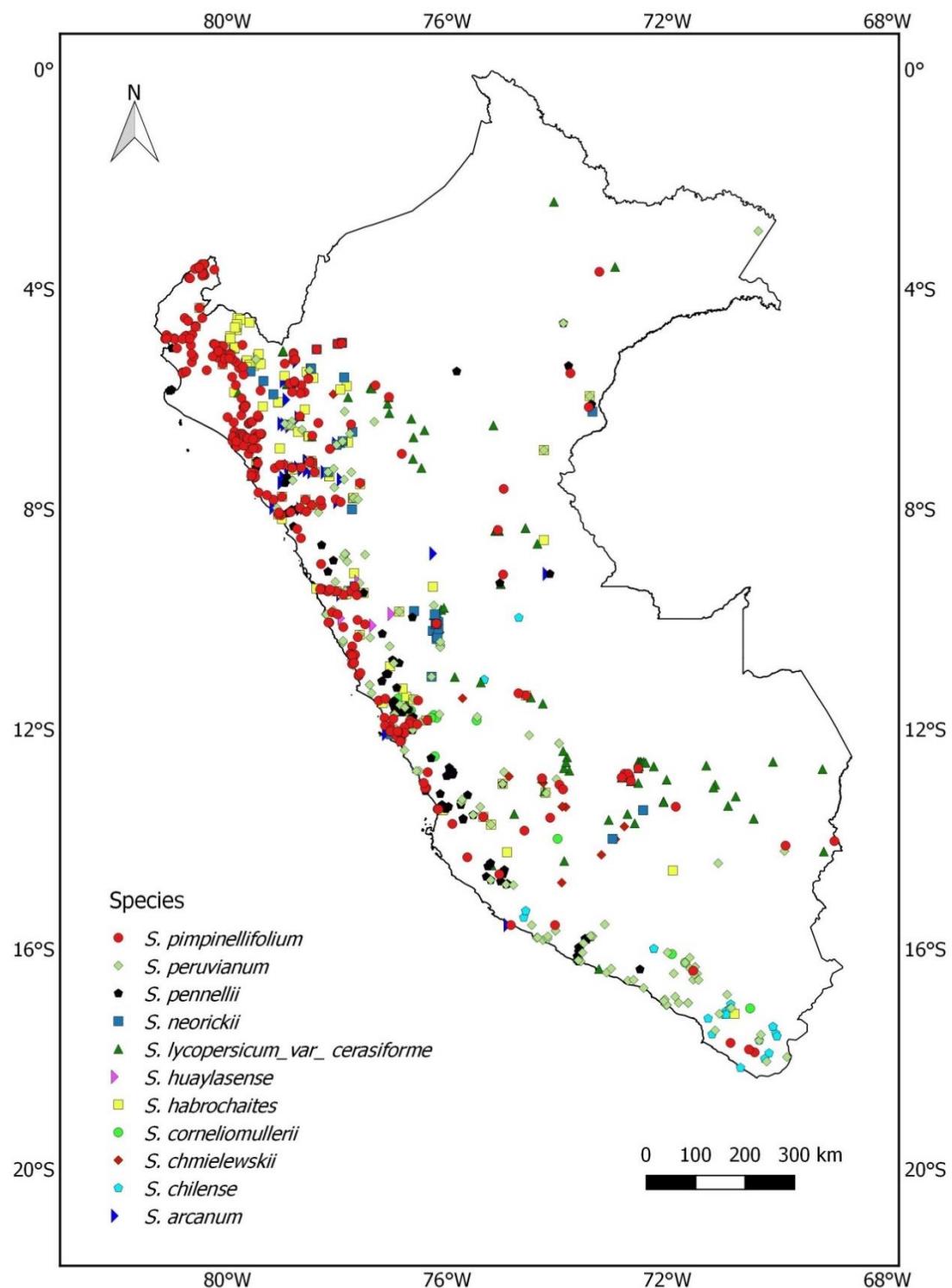


Figure 1. Wild tomato species distributed in Peru based on their herbaria and germplasm collection data.

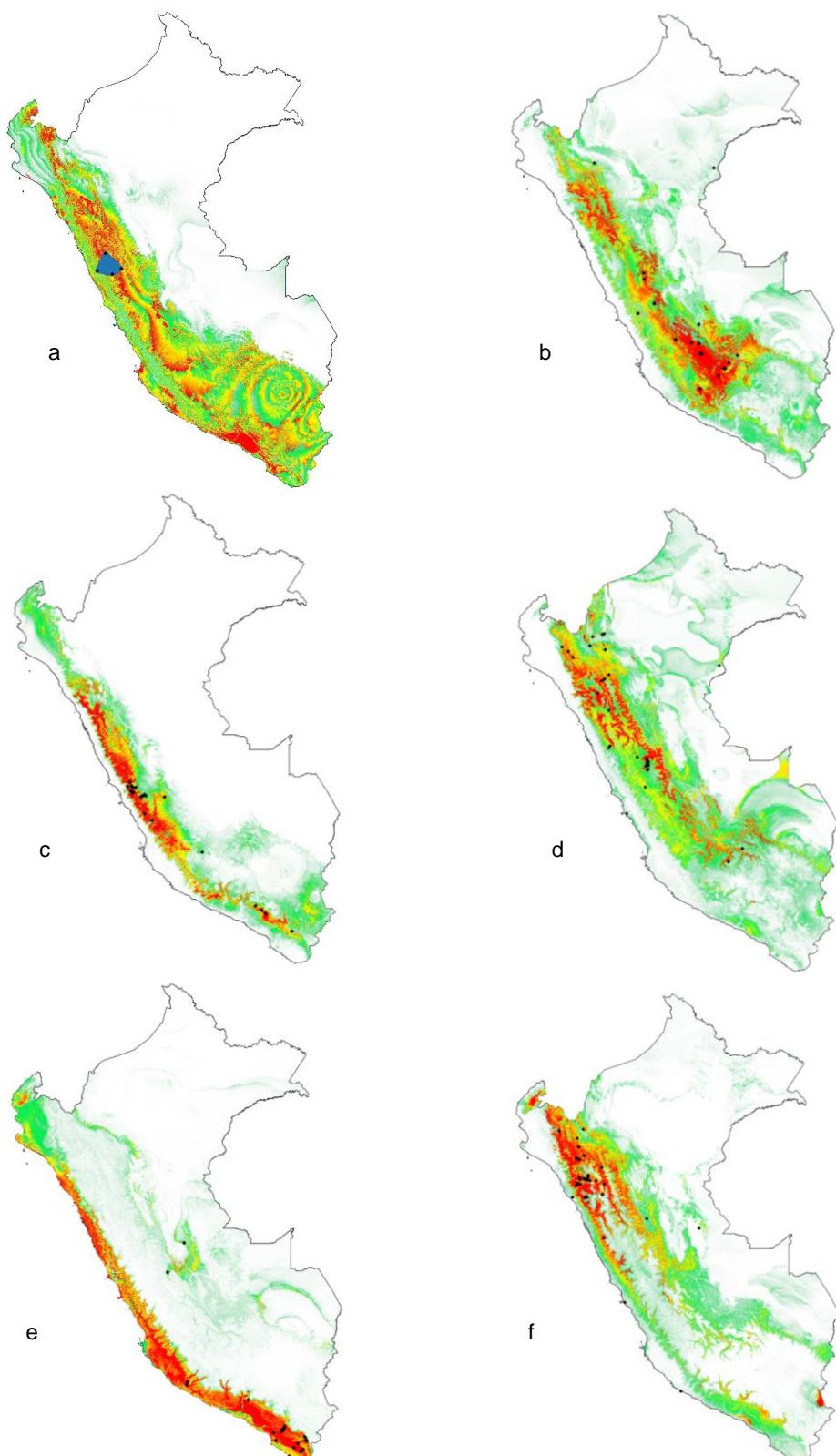


Figure 2. Species distribution models along with collection coordinates for tomato's wild relatives with higher priority for collection for INIA. a) *Solanum huaylasense*, b) *S chmielewskii*, c) *S. corneliomullerii*, d) *S. neorickii*, e) *S. chilense*, f) *S. arcanum*. Red areas correspond to sites where the species is most likely to occur.

PERUVIAN TOMATO GAP ANALYSIS

Julio - Diciembre 2019

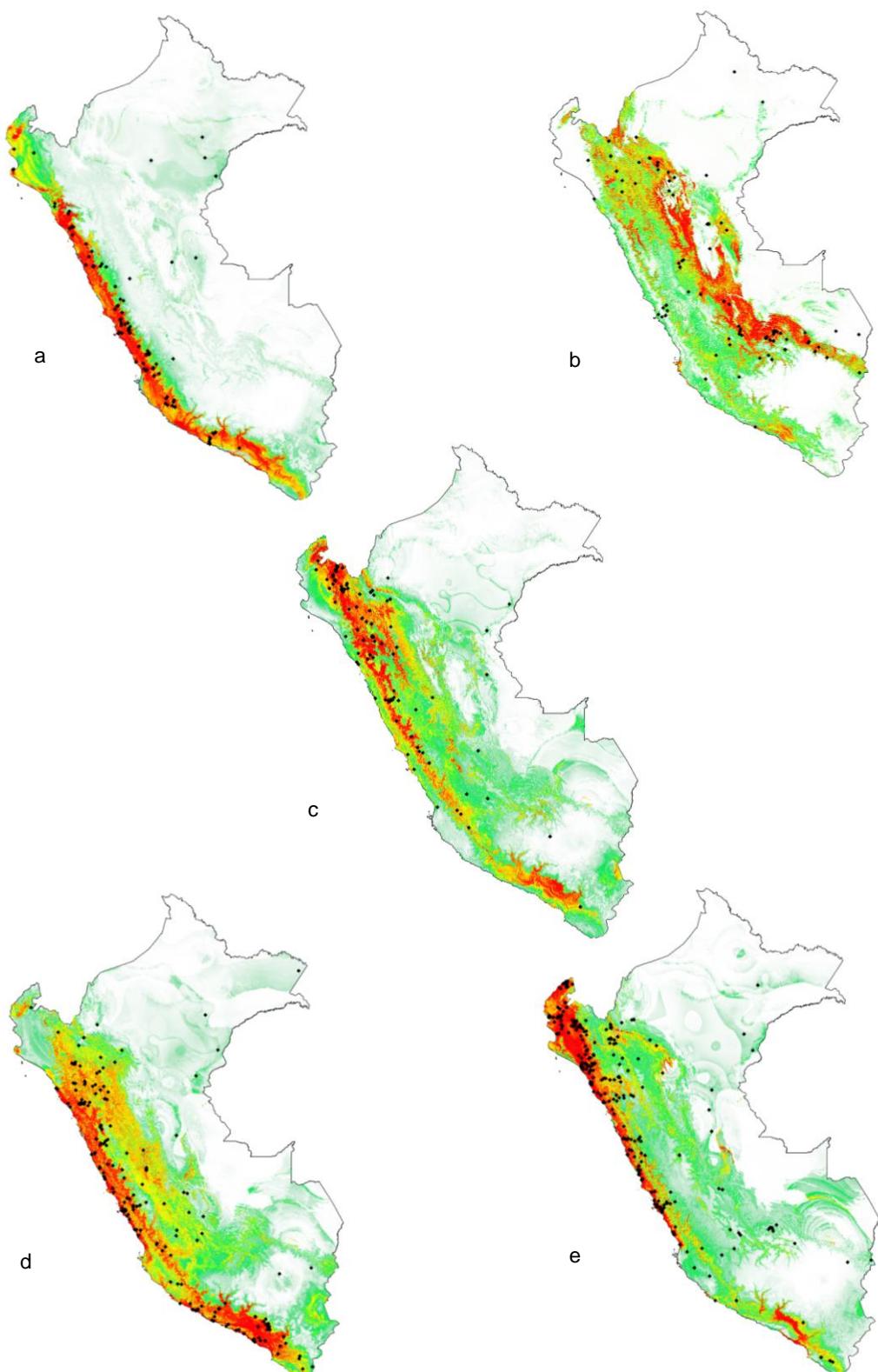


Figure 3. Species distribution models along with collection coordinates for tomato's wild relatives with medium priority for collection for INIA. a) *Solanum pennellii*, b) *S. lycopersicum* var. *ceraciforme*, c) *S. habrochaites*, d) *S. peruvianum*, e) *S. pimpinellifolium*. Red areas correspond to sites where the species is most likely to occur.

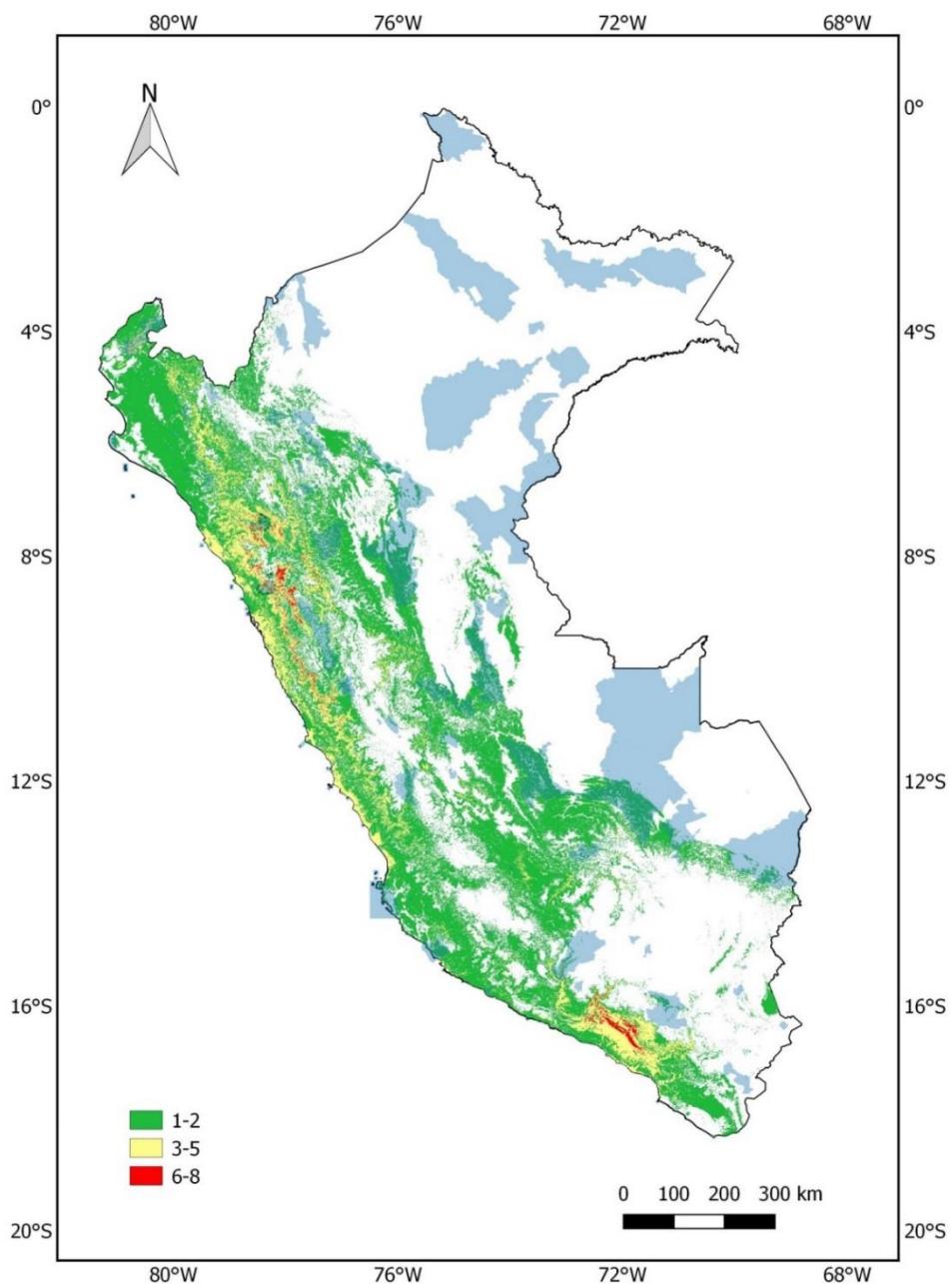


Figure 4. Final prioritization of collecting areas for the Peruvian wild tomato species, where red areas correspond to sites where there is higher concentration of wild relatives (numbers indicate the number of occurring wild relatives). Light blue areas correspond to the Peruvian National System of Protected Areas.

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